



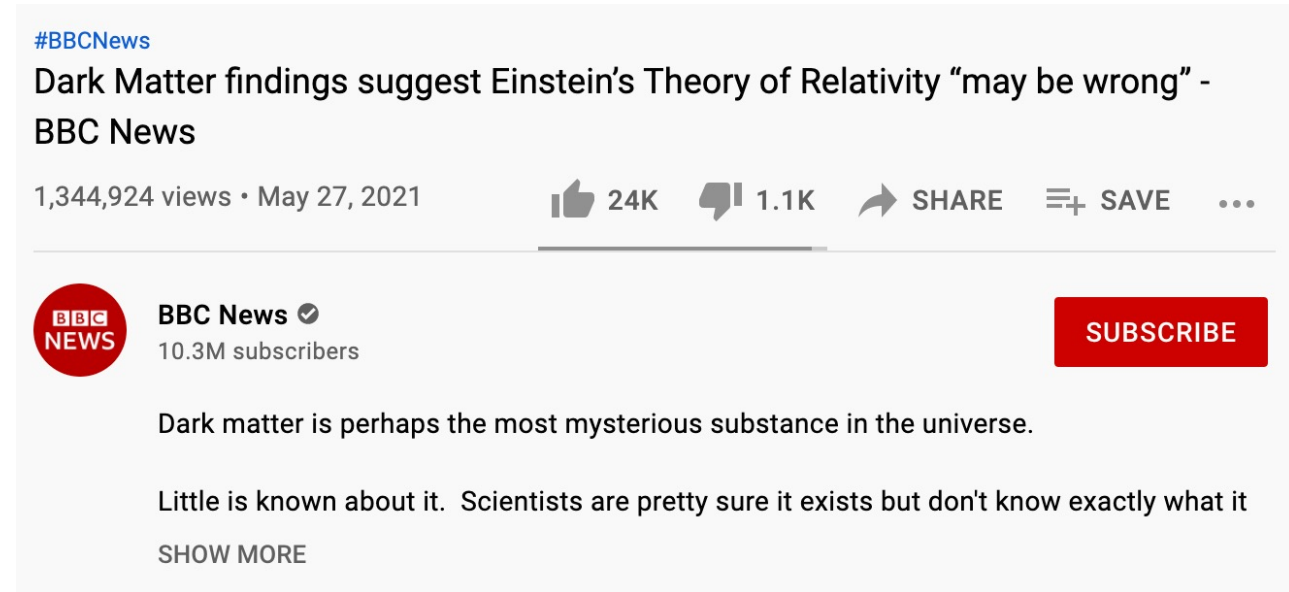
DES Year 3 Results: Cosmology from weak lensing and galaxy clustering

Niall MacCrann (DAMTP, Cambridge),
on behalf of the DES Collaboration



What I'll talk about

- A quick large-scale structure cosmology intro
- A quick Dark Energy Survey intro
- A bit about weak gravitational lensing
- The DES Year 3 “3x2pt” analysis and challenges
- Cosmological constraints
- Was Einstein wrong? (not yet, clickbait is real...)





Large scale structure cosmology

- We have a concordance cosmological model, (flat) Λ CDM.
- Observationally solid, but what is Λ ? (and what is CDM!!?)
- Strong constraints from mature geometrical probes:
 - Standard rulers (BAO in CMB/galaxies)
 - Standard candles (SN1a)
- Largely depend on background quantities e.g. Average energy densities and expansion rate of the Universe.





Large scale structure cosmology

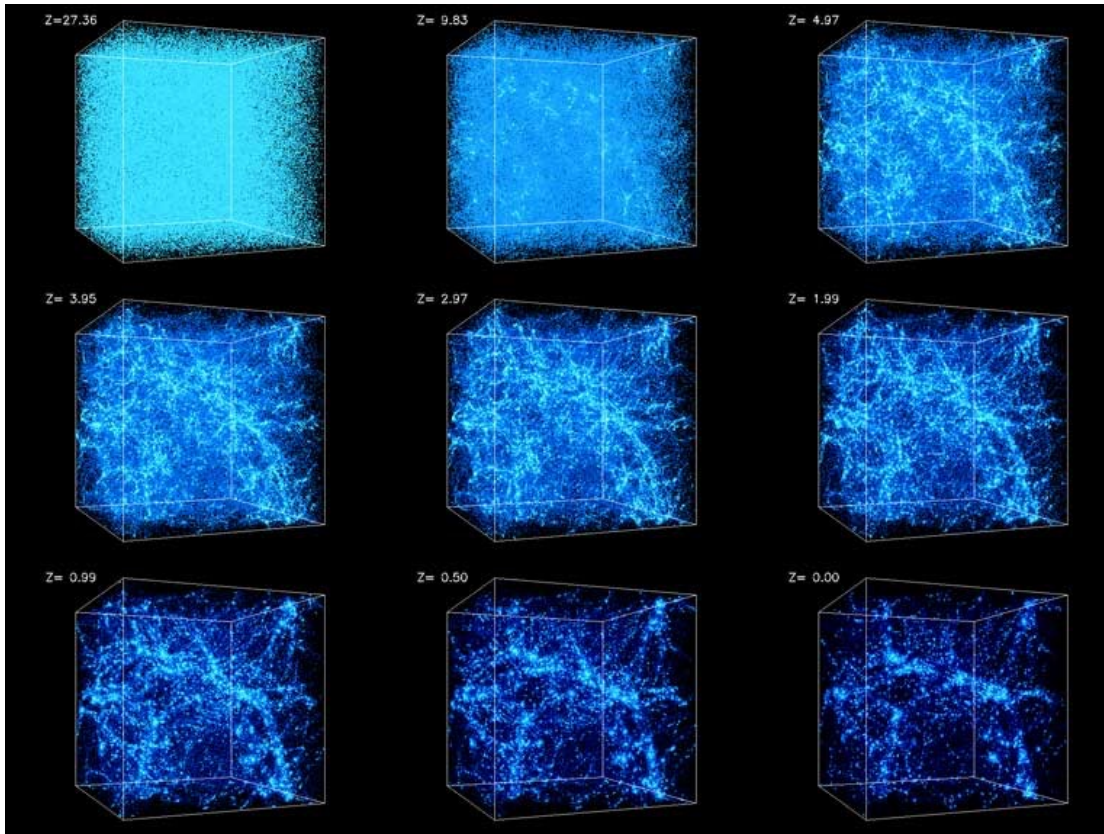
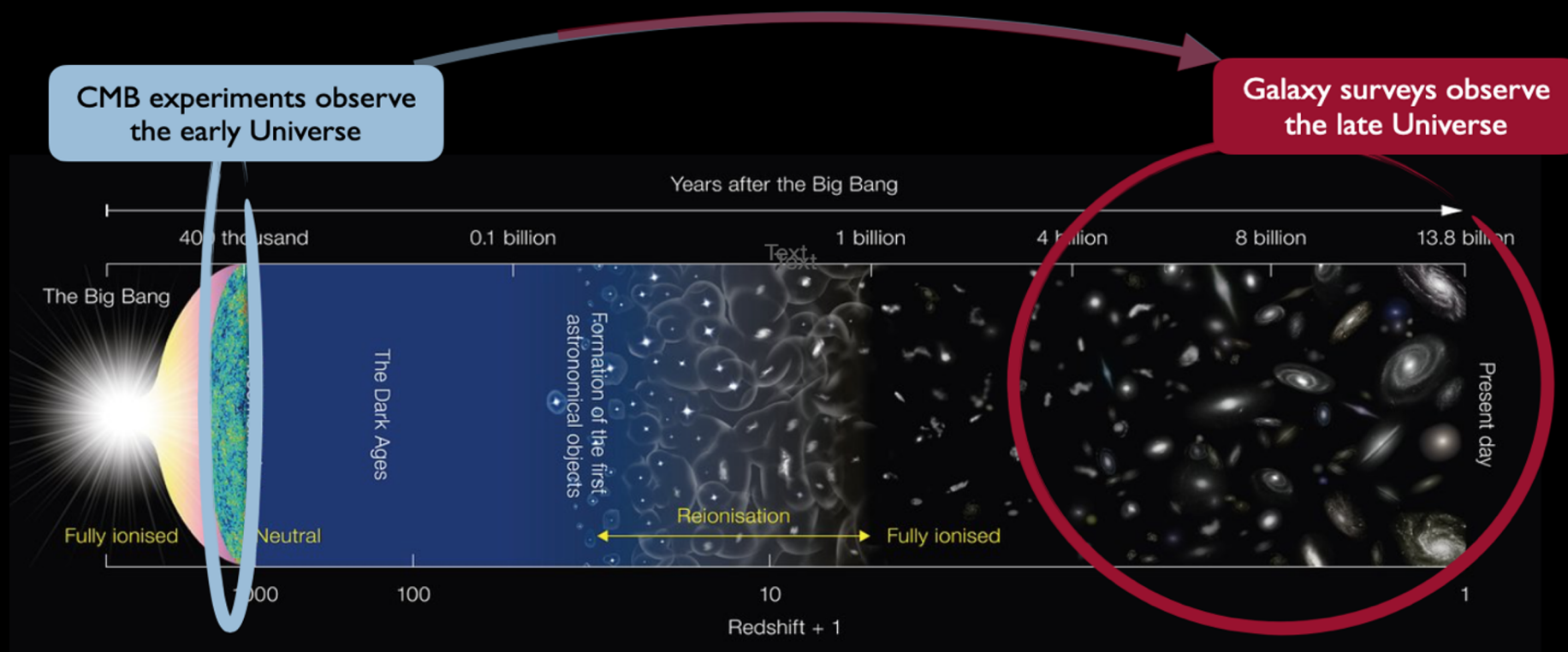


Image credit: Andrey Kravtsov and Anatoly Klypin

- What about the perturbations?
- The statistics of the density field can tell us about the **growth of structure** — complementary to geometrical information.



Testing Λ CDM: Is the late time clustering compatible with the Λ CDM prediction assuming initial conditions from the CMB?

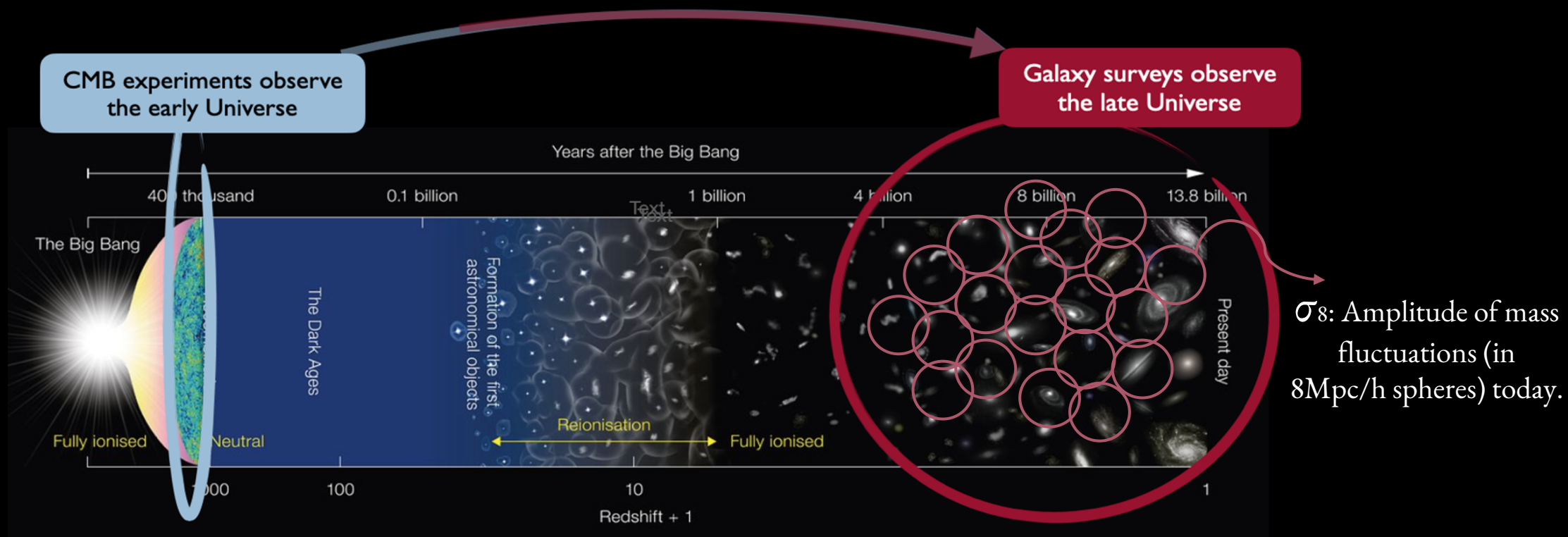


A_s : Amplitude of primordial scalar density fluctuations.

Image credit: NAOJ



Testing Λ CDM: Is the late time clustering compatible with the Λ CDM prediction assuming initial conditions from the CMB?



A_s : Amplitude of primordial scalar density fluctuations.

Image credit: NAOJ

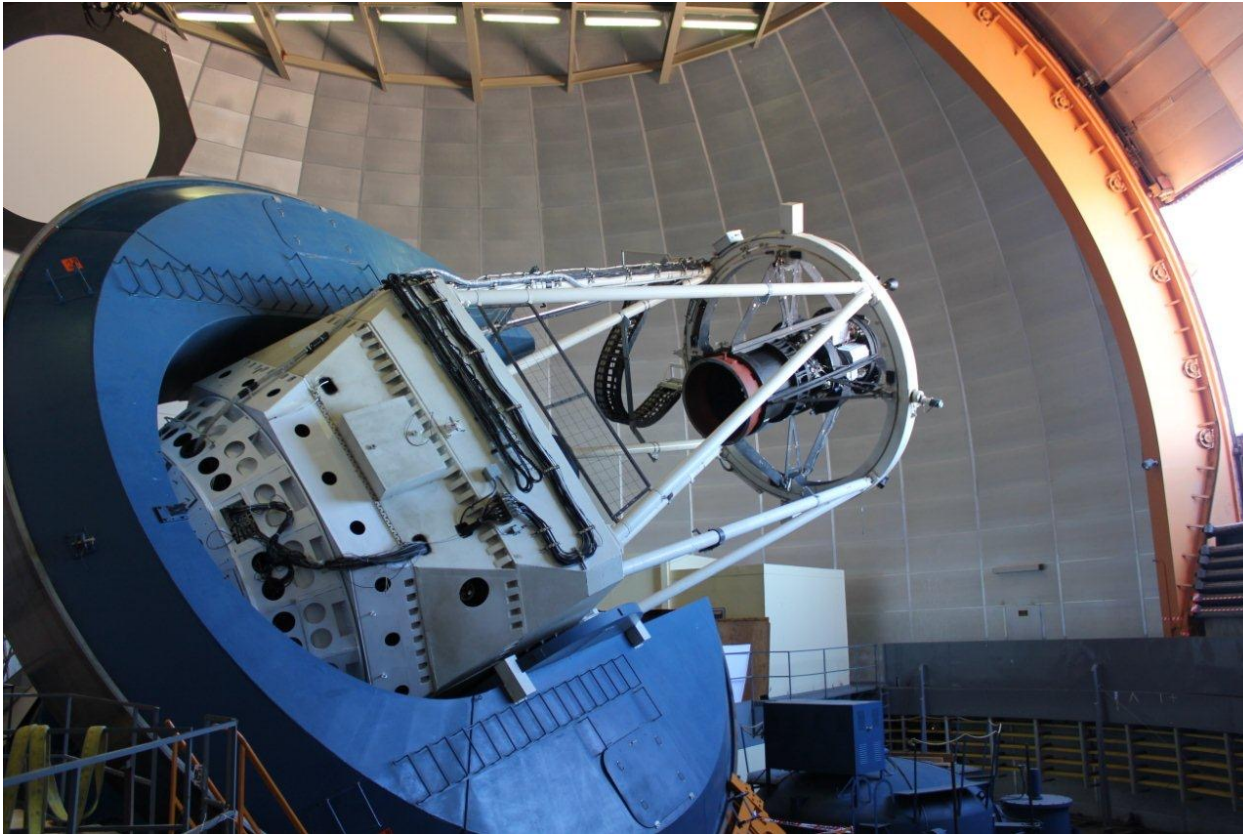


What I'll talk about

- A quick large-scale structure cosmology intro
- A quick Dark Energy Survey intro
- A bit about weak gravitational lensing
- The DES Year 3 “3x2pt” analysis and challenges
- Cosmological constraints



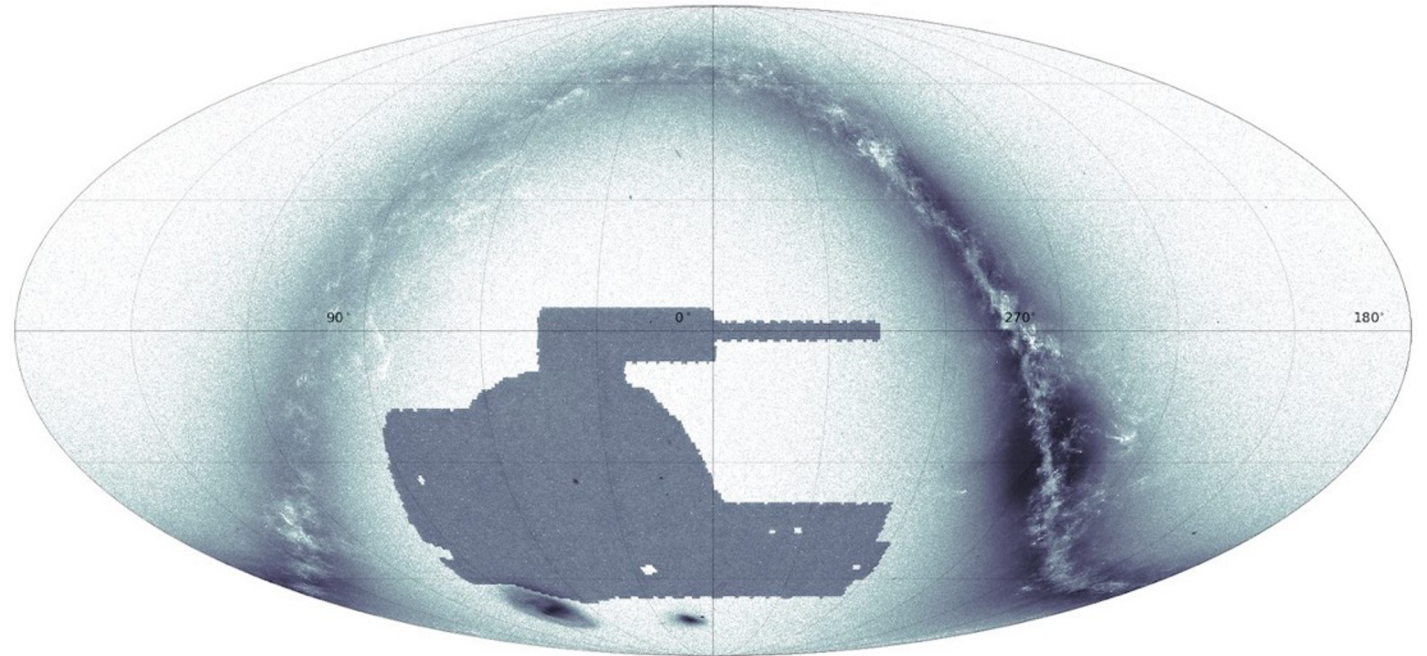
The Dark Energy Survey (DES)





The Dark Energy Survey (DES)

- 570 Megapixel camera for the Blanco 4m telescope in Chile.
- Full survey 2013-2019 (Year 3 2013-16).
- Wide field: 5000 sq. deg. in 5 bands. ~ 23 magnitude.
- DES Y3: Positions and shapes of $> 100\text{M}$ galaxies.

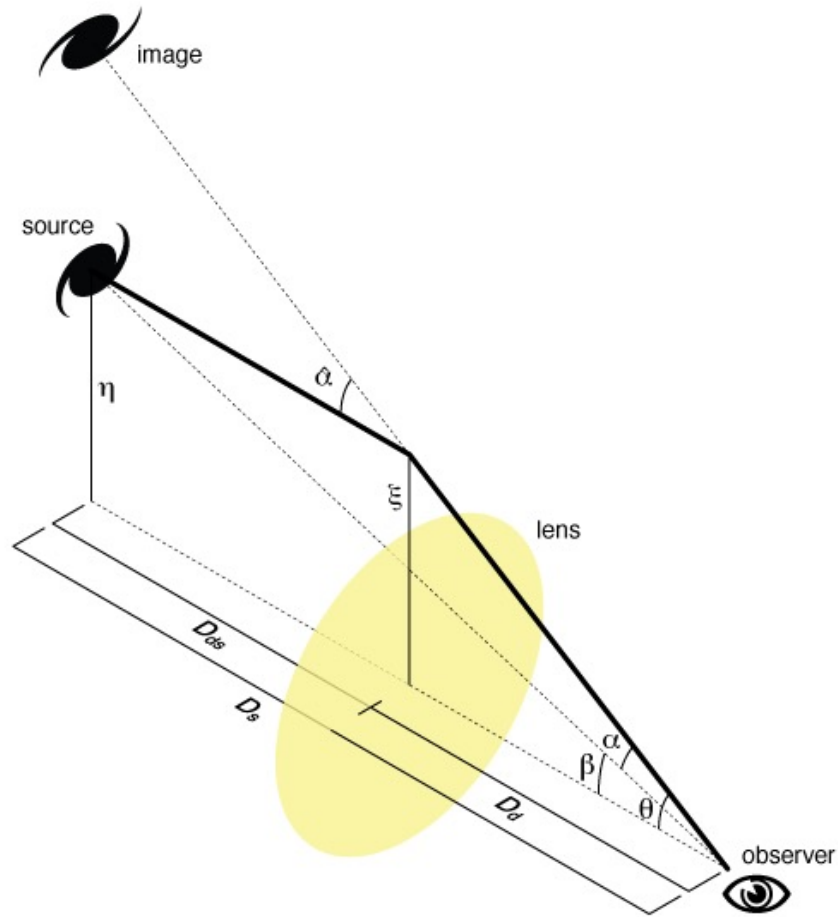




What I'll talk about

- A quick large-scale structure cosmology intro
- A quick Dark Energy Survey intro
- A bit about weak gravitational lensing
- The DES Year 3 “3x2pt” analysis and challenges
- Cosmological constraints

Weak Gravitational Lensing



$$\hat{\alpha} = \frac{4GM}{c^2\xi}$$

<- For a point mass

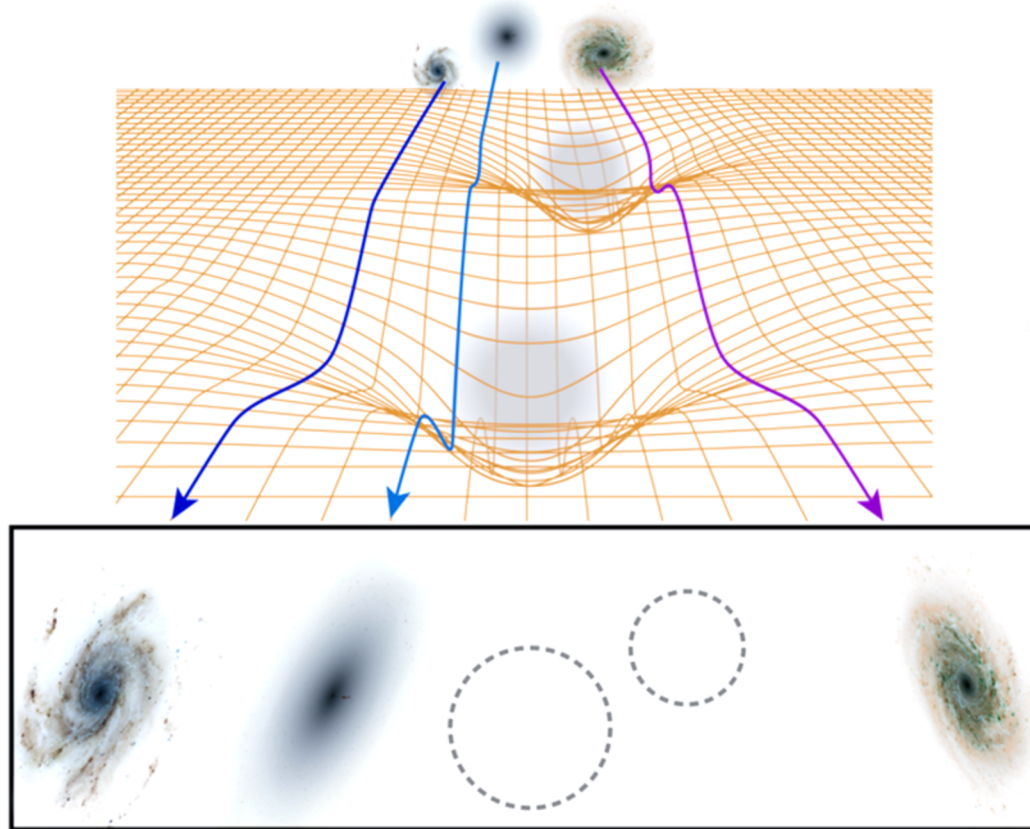
$$\vec{\alpha}(\vec{\theta}) = \vec{\nabla}\psi(\vec{\theta})$$

<- In general

$$\psi(\vec{\theta}) = \frac{2D_{ds}}{D_d D_s c^2} \int \Phi(D_d \vec{\theta}, z) dz$$



Weak Gravitational Lensing



To first order, unlensed coordinates are transformed to the lensed coordinates according to the lensing Jacobian

$$A_{ij} = \delta_{ij} - \frac{\partial \psi}{\partial \theta_i \partial \theta_j} = \begin{bmatrix} 1 - \kappa - \gamma_1 & \gamma_2 \\ \gamma_2 & 1 - \kappa + \gamma_1 \end{bmatrix}$$

Galaxy ellipticities transform as

$$\epsilon \approx \epsilon_s + \gamma$$

Galaxies are randomly oriented (before lensing) so:

$$\langle \epsilon \rangle = \langle \epsilon_s \rangle + \langle \gamma \rangle = \langle \gamma \rangle$$

APS/[Alan Stonebraker](#); galaxy images from STScI/AURA, NASA, ESA, and the Hubble Heritage Team



Weak Gravitational Lensing

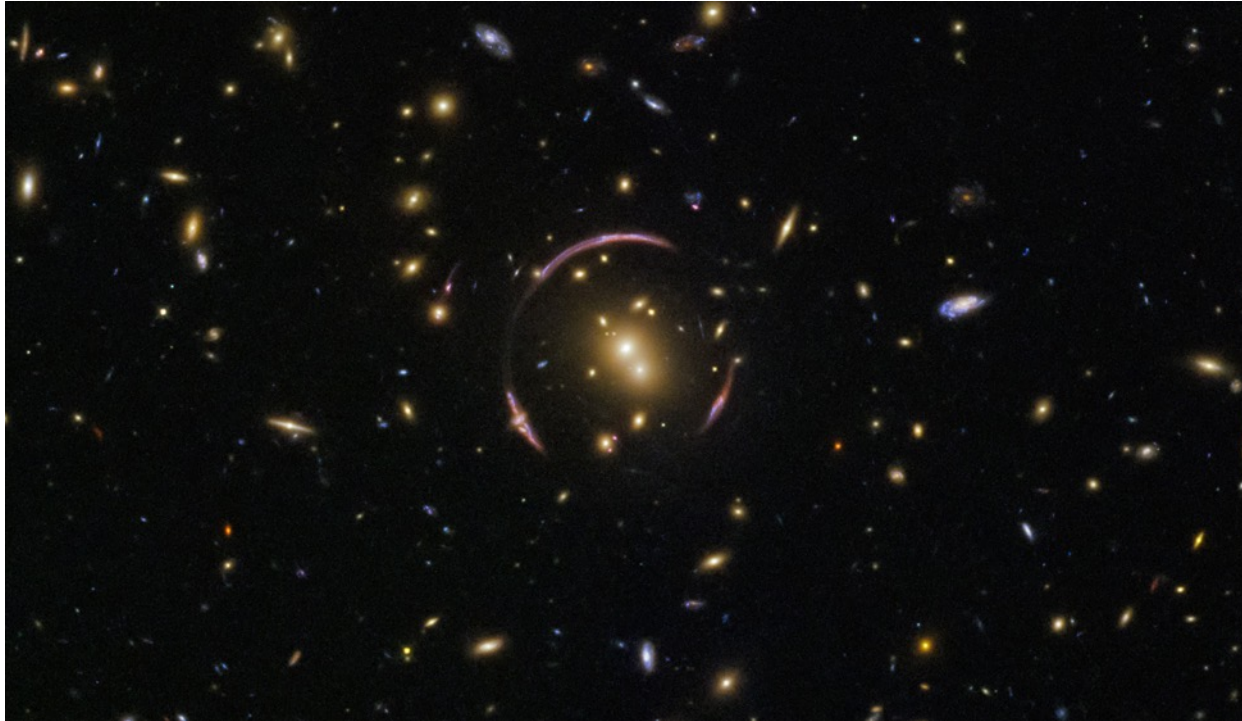


Image credit: ESA/Hubble
& NASA

Strong gravitational lensing:

- Prettier
- Much rarer!



What I'll talk about

- A quick large-scale structure cosmology intro
- A quick Dark Energy Survey intro
- A bit about weak gravitational lensing
- The DES Year 3 “3x2pt” analysis and challenges
- Cosmological constraints



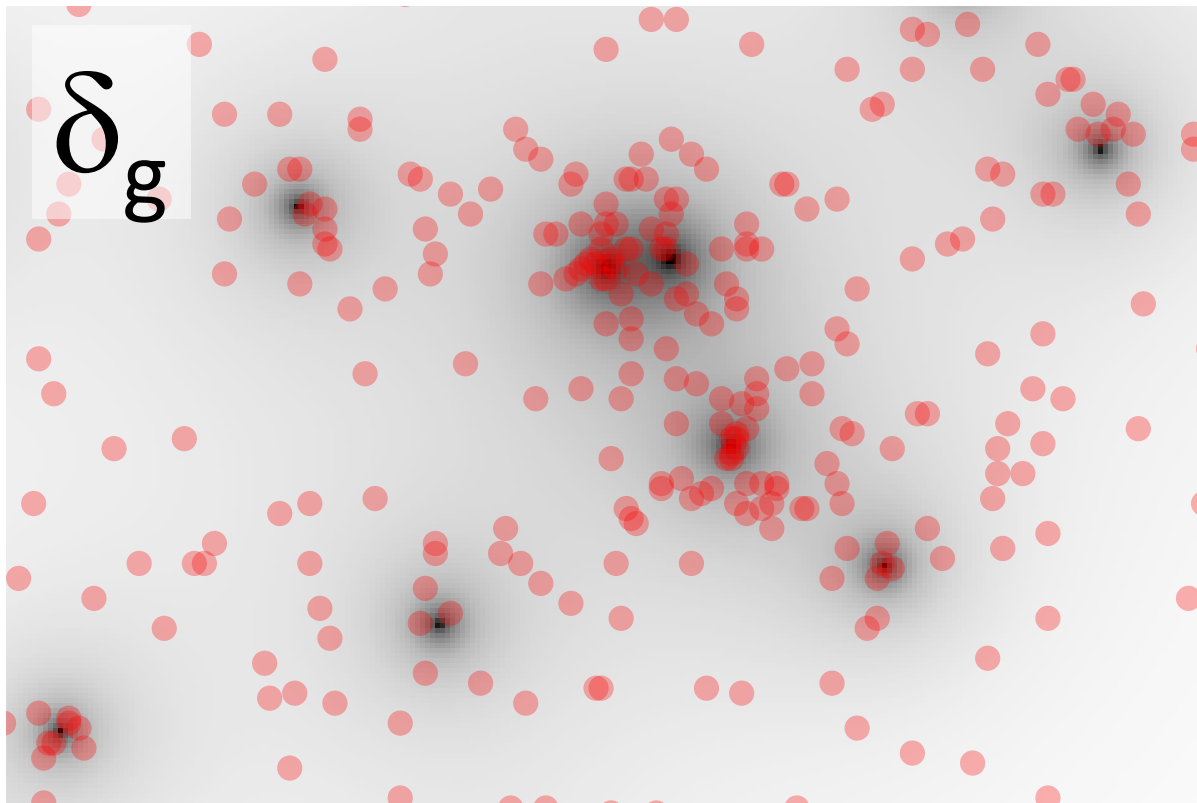
The Observables



We want to access the matter perturbations δ_m
(because their e.g. 2-point statistics are
sensitive to cosmology)



The Observables



We use two observables to do this:

1. δ_g : Counts of galaxies:

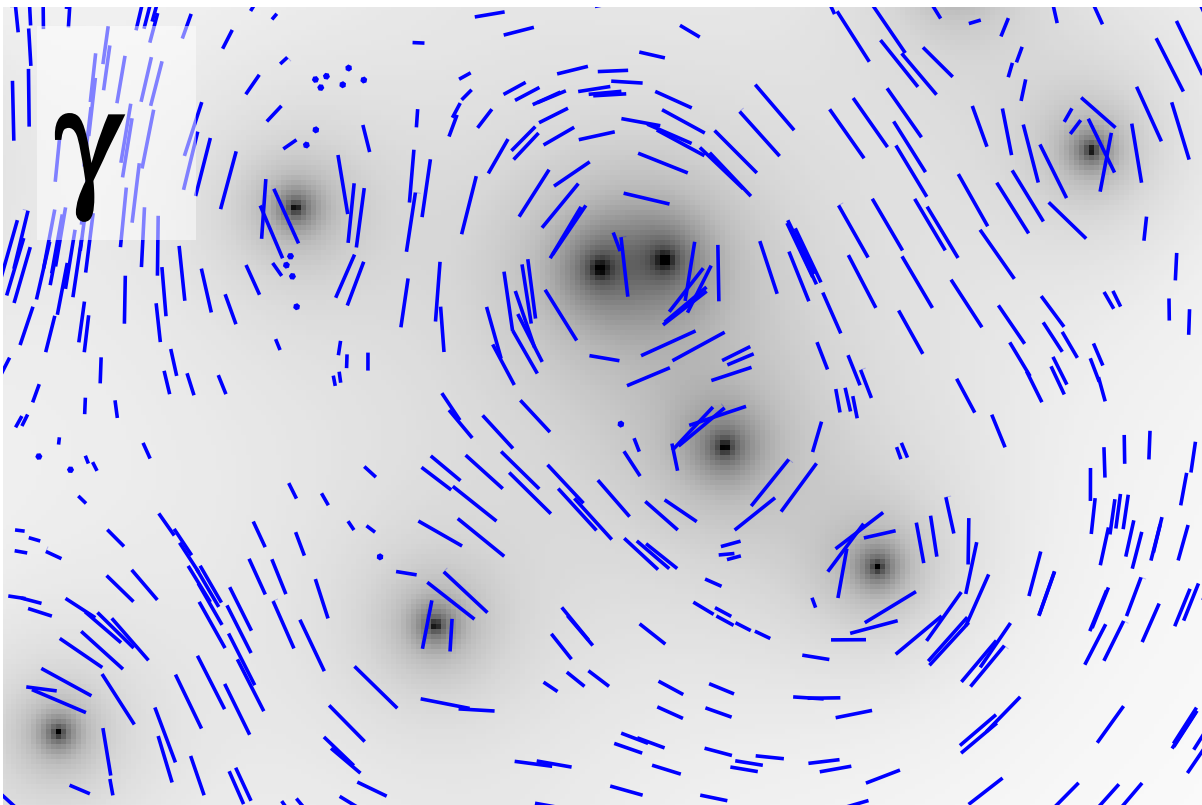
$$\langle \delta_g \delta_g \rangle = b^2 \langle \delta_m \delta_m \rangle$$

b is the **galaxy bias** and is unknown...

“galaxy clustering”



The Observables



We use two observables to do this:

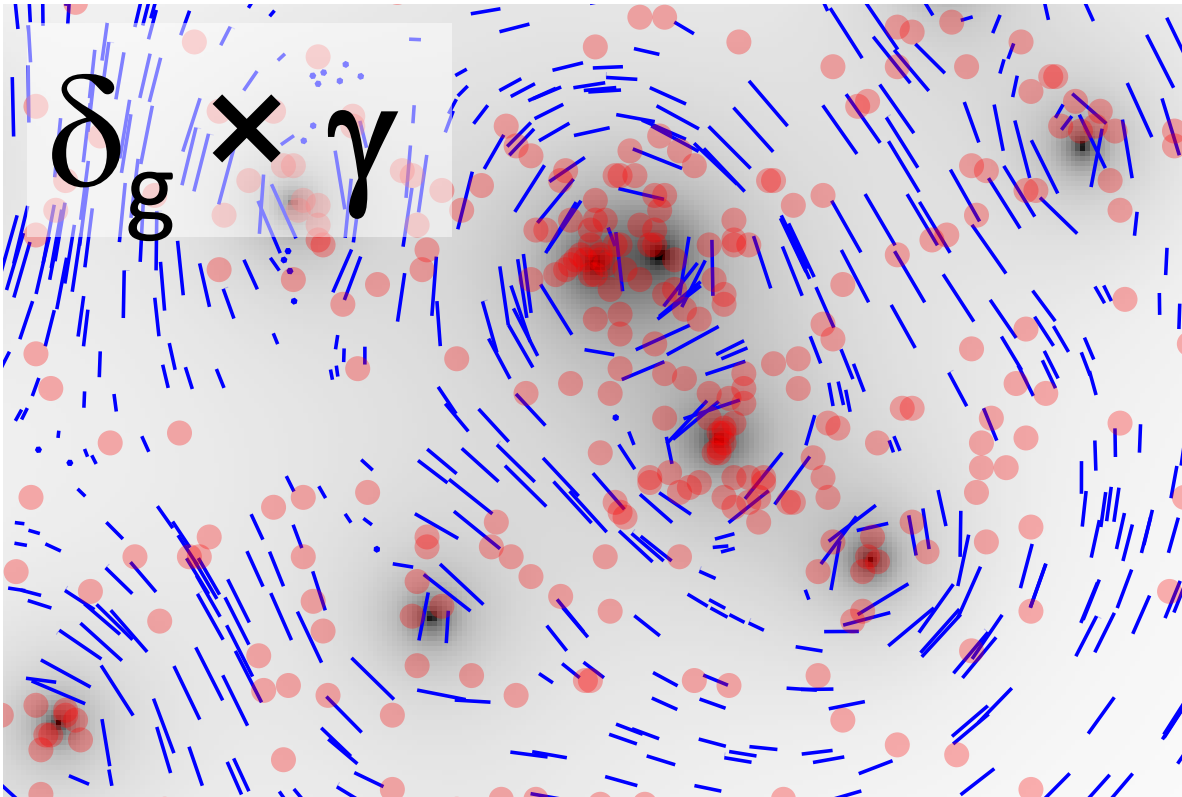
2. Weak lensing shear, γ :
 - Produces **coherent galaxy ellipticities**.
 - Depends directly on projected δ_m :

$$\langle \gamma \gamma \rangle \sim \langle \delta_m \delta_m \rangle$$

“cosmic shear”



The Observables



We can also use the cross-correlation:

$$\langle \delta_g \gamma \rangle \sim b \langle \delta_m \delta_m \rangle$$

Also sensitive to galaxy bias

“galaxy-galaxy lensing”

“Lens galaxies”

“Source galaxies”



The Observables

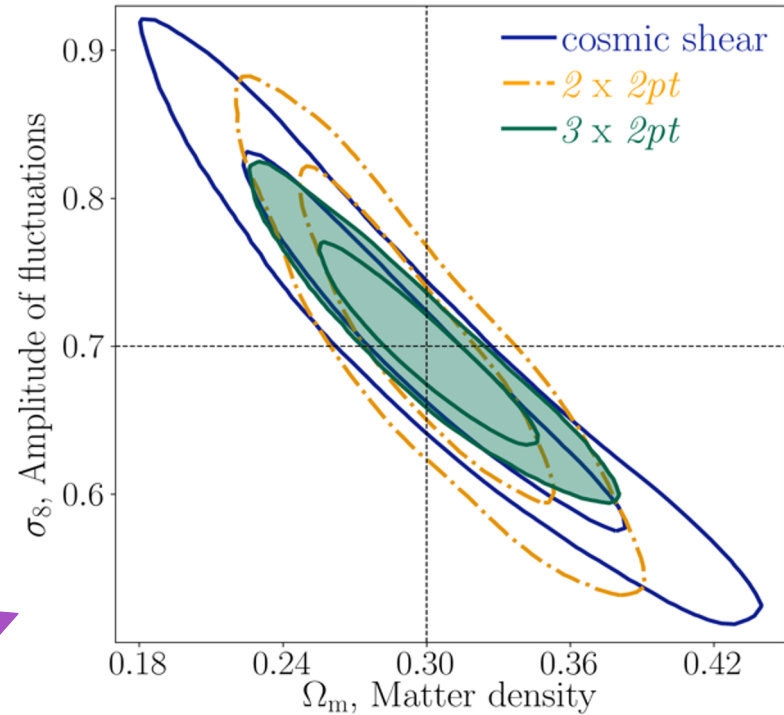
Summarizing, we use...

1. Galaxy number density field δ_g
2. The weak lensing shear field γ

i) $\langle \gamma \gamma \rangle \sim \langle \delta_m \delta_m \rangle$

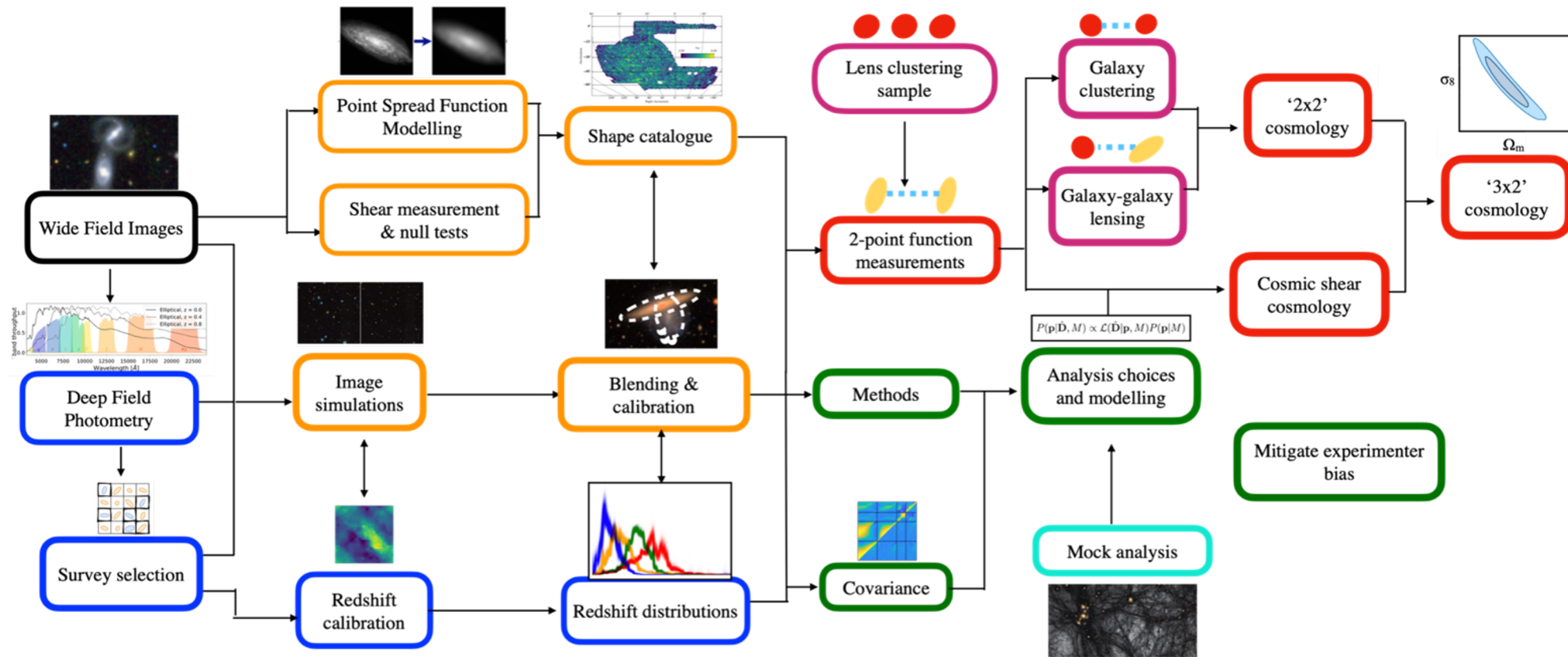
ii) $\langle \delta_g \delta_g \rangle = b^2 \langle \delta_m \delta_m \rangle$

iii) $\langle \delta_g \gamma \rangle \sim b \langle \delta_m \delta_m \rangle$



(where $\langle \rangle$ means 2 point correlation function or power spectrum)

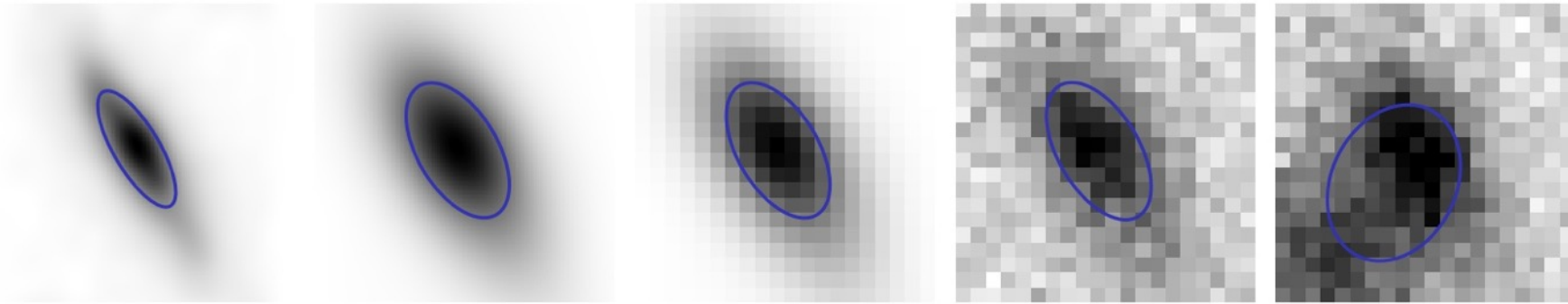
The DES Year 3 analysis



LCDM — WL+LSS — Redshifts — Shapes — Clustering — Simulations — Theory — Results



The Challenges: Shear estimation



Galaxy images have:

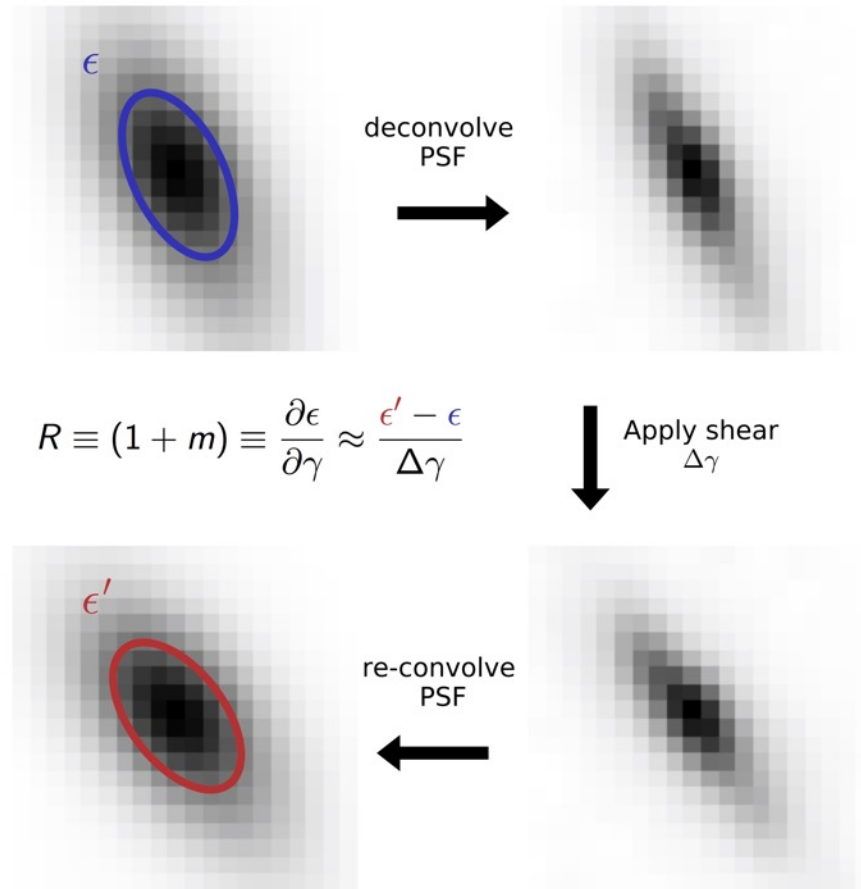
- Complex morphologies (ellipticity not uniquely defined)
- Blurring due to the atmosphere / telescope optics
- Noise
- Blending

Shear estimation biases parameterized as

$$\langle \epsilon^{\text{obs}} \rangle = (1 + m) \langle \gamma \rangle + c$$



The Challenges: Shear estimation



Metacalibration

- A method for calibrating the shear estimate without requiring complex calibration simulations (Huff & Mandelbaum 2017, Sheldon & Huff 2017).
- Generalized to full scenes in Sheldon, Becker, NM et al. 2020.
- But for blends between galaxies at different redshifts, calibration simulations still required.

The image is split into two vertical panels. The left panel shows a sparse field of galaxies with various colors (blue, green, orange, red) against a black background. The right panel shows a similar field but with more prominent, bright blue and cyan galaxies, suggesting a different selection or simulation. Both panels contain numerous smaller, fainter galaxies of various colors.

Which is real vs simulated?

Simulate galaxy images in multiple photometric bands and apply the same measurement pipeline

MacCrann+2021



Which is real vs simulated?

simulated

Simulate galaxy images in multiple photometric bands and apply the same measurement pipeline

MacCrann+2021

real

Calibrate shear biases with image simulations

$$\bar{\epsilon}^{\text{obs}} = (1 + m) \bar{\gamma} + c$$

observed ellipticity *multiplicative error* *lensing shear* *additive error*

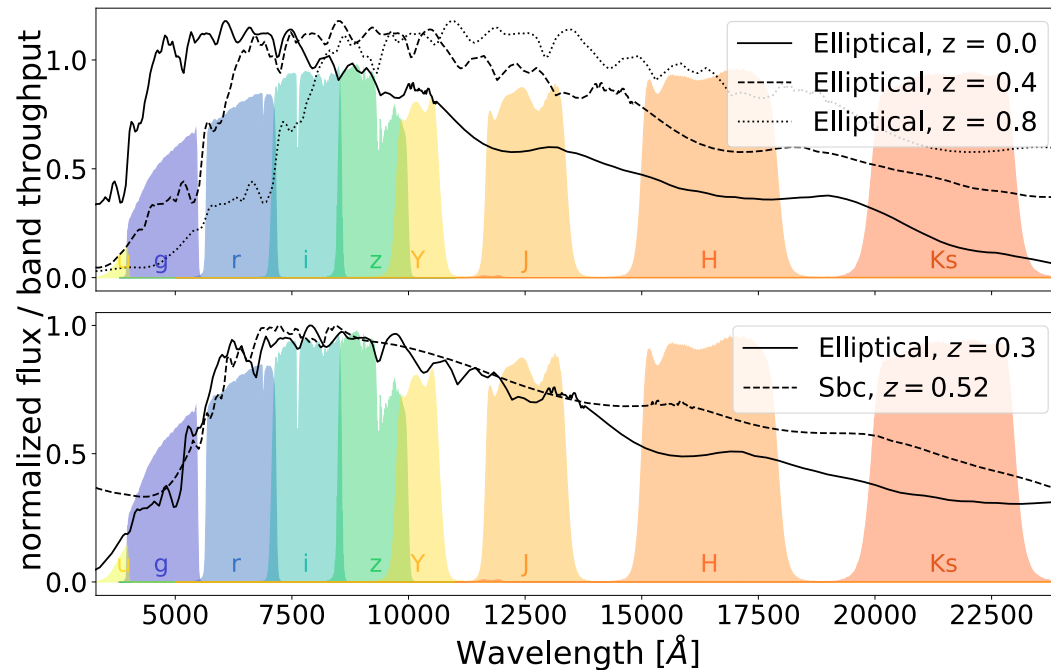
- Few percent multiplicative biases **due to blending** (-1.5 to -4% depending on redshift bin)
- Joint impact of blending on shear and photo-z characterized by **effective redshift distribution**

$$\bar{\epsilon}^{\text{obs}} = \int dz n_{\gamma}(z) \gamma^{\text{true}}(z) + c + \text{noise}$$

MacCrann+2021



The Challenges: Redshift estimation



Buchs et al. 2019

- Accurate redshift distributions $n(z)$ required for theoretical predictions.
- Imaging surveys have a limited number of bands.
- Redshifts estimated from this very crude spectrum.
- Degeneracy between galaxy type and redshift



The Challenges: Redshift estimation

- In Year 3 we used a novel “SOMPZ” method (see Buchs et al. 2019, Alarcon et al. 2019, Myles et al. 2021).
- DES also has a smaller ($\sim 30 \text{ deg}^2$) **deep survey** which has overlap with near infrared data (J, H, Ks bands – see Hartley, Choi et al. 2021). In this area, highly accurate and precise redshift estimation is possible. One can leverage this information via

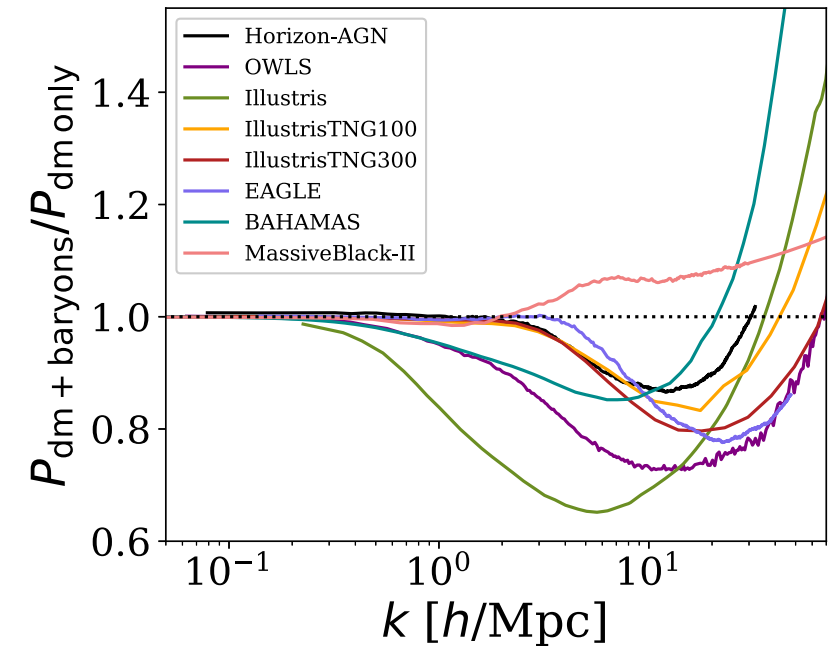
$$\underbrace{P(z|\text{wide flux})}_{\text{Accurate redshifts in wide fields}} = \int dz \underbrace{P(\text{wide flux}|\text{deep flux})}_{\text{Survey transfer function}} \underbrace{P(z|\text{deep flux})}_{\text{Precise redshifts in deep fields}}$$



The Challenges: Theoretical Predictions

- Weak lensing observables are sensitive to the small-scale matter power spectrum ($k > 1 \text{ h/Mpc}$). Galactic astrophysics (“Baryonic effects”), affects the matter distribution here, much harder to simulate than gravity-only sims.
- Nonlinear galaxy bias and intrinsic alignments becoming important:

$$\delta_g = b_1 \delta_m + b_2^2 \delta_m^2 + \text{other } O(\delta_m^2) + O(\delta_m^3) + \dots$$

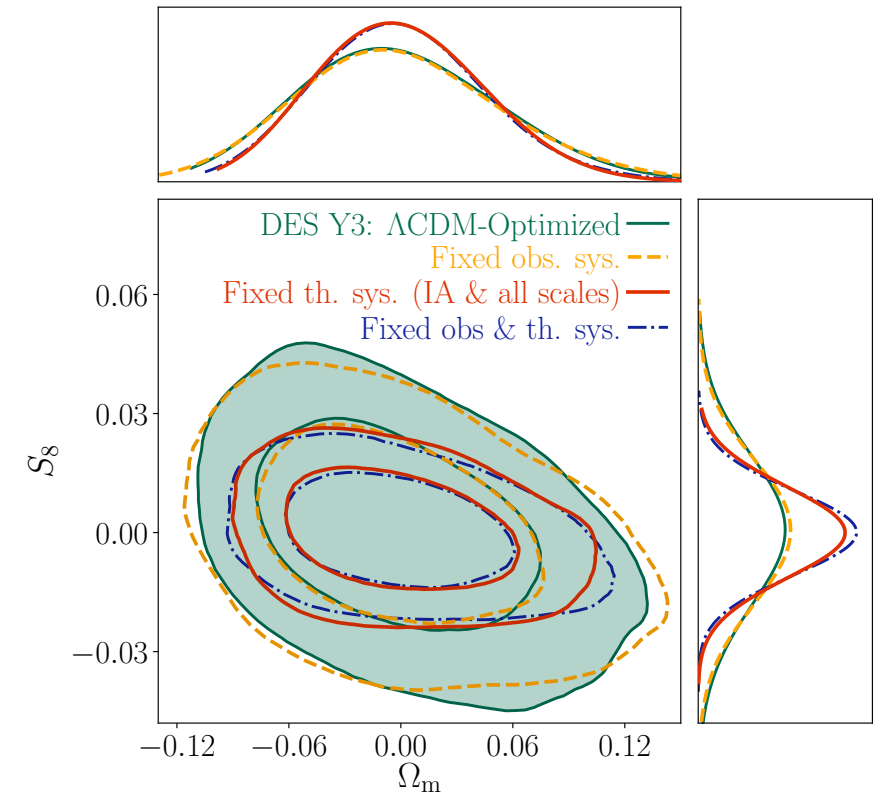


Chisari et al. 2019



The Challenges: Theory Predictions

- We largely address these issues by throwing away information!
 - Conservative scale cuts (see Krause et al. 2021 and Secco & Samuroff 2021)
 - Extra nuisance parameters (see Pandey et al. 2019 for nonlinear bias model)
 - Analytic marginalization schemes (see NM et al. 2019 for “point-mass” marginalization)
- Theoretical uncertainties were our limiting systematics for DES Y3 cosmic shear



Amon et al. 2021



What I'll talk about

- A quick large-scale structure cosmology intro
- A quick Dark Energy Survey intro
- A bit about weak gravitational lensing
- The DES Year 3 “3x2pt” analysis and challenges
- Cosmological constraints
(<https://arxiv.org/abs/2105.13549>)

Drumroll....

The Results: Internal consistency

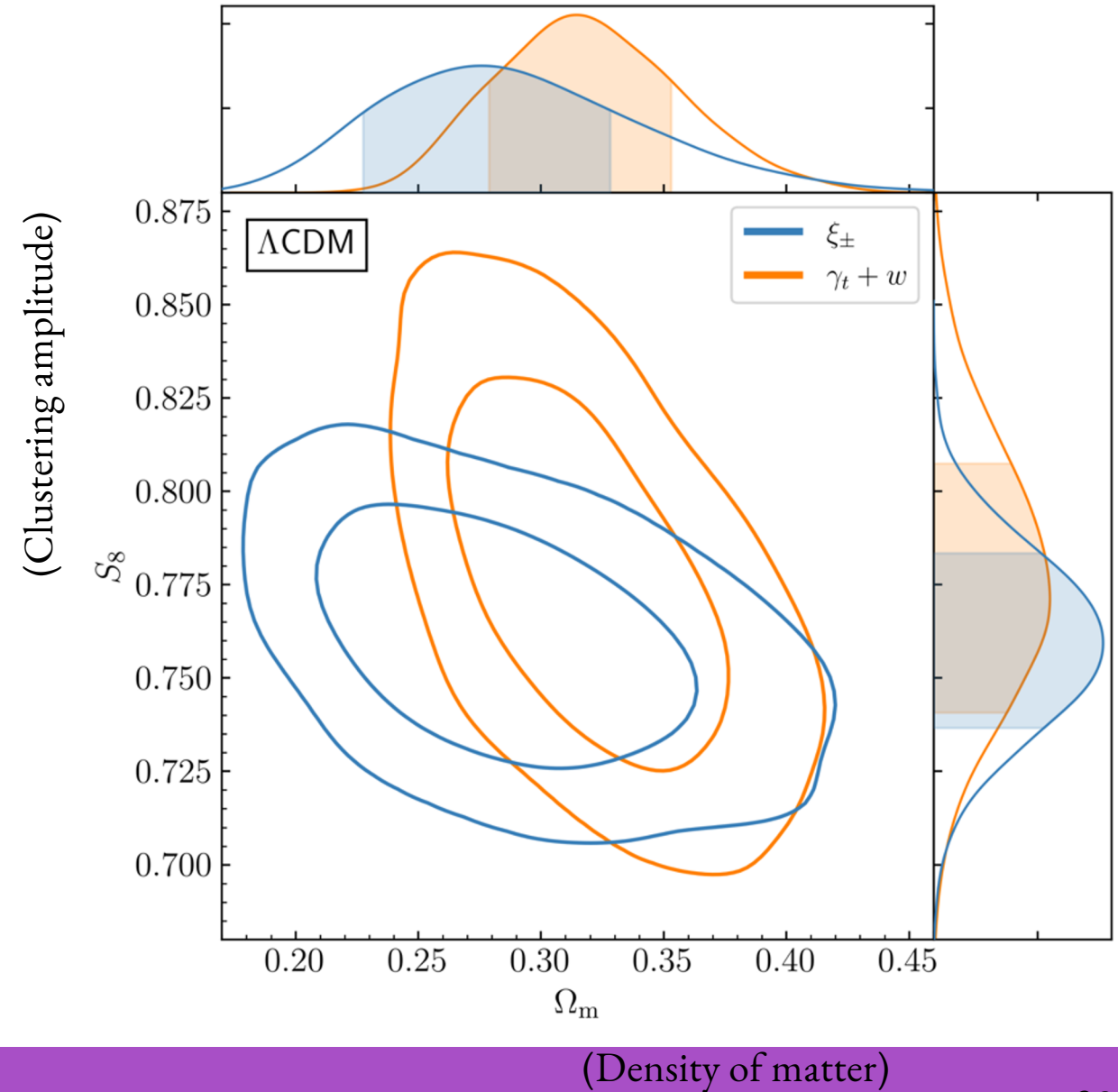
Two correlated cosmological probes:

1. **Cosmic shear** (blue)
2. **Galaxy clustering and tangential shear** (orange)

We find consistency between them.

Cosmic shear most sensitive to clustering amplitude.

Galaxy clustering and tangential shear more sensitive to total matter density.



The Results: 3x2pt

- Combine these into the **3x2pt** probe of large-scale structure.
- Factor of 2 improvement in constraining power w.r.t. DES Year 1.

In Λ CDM:

$$S_8 = 0.776^{+0.017}_{-0.017} \quad (0.776)$$

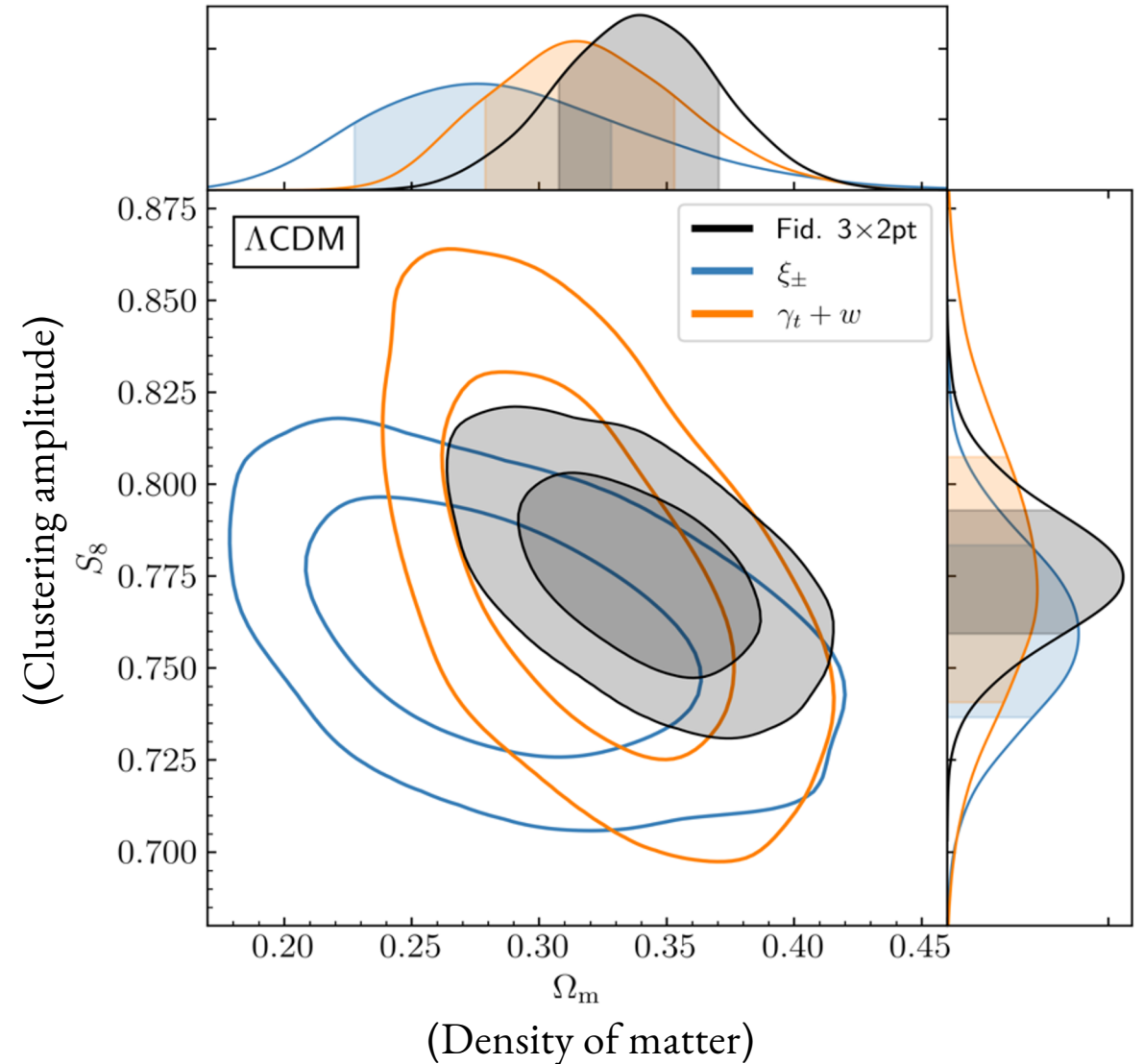
$$\Omega_m = 0.339^{+0.032}_{-0.031} \quad (0.372)$$

$$\sigma_8 = 0.733^{+0.039}_{-0.049} \quad (0.696)$$

In w CDM:

$$\Omega_m = 0.352^{+0.035}_{-0.041} \quad (0.339)$$

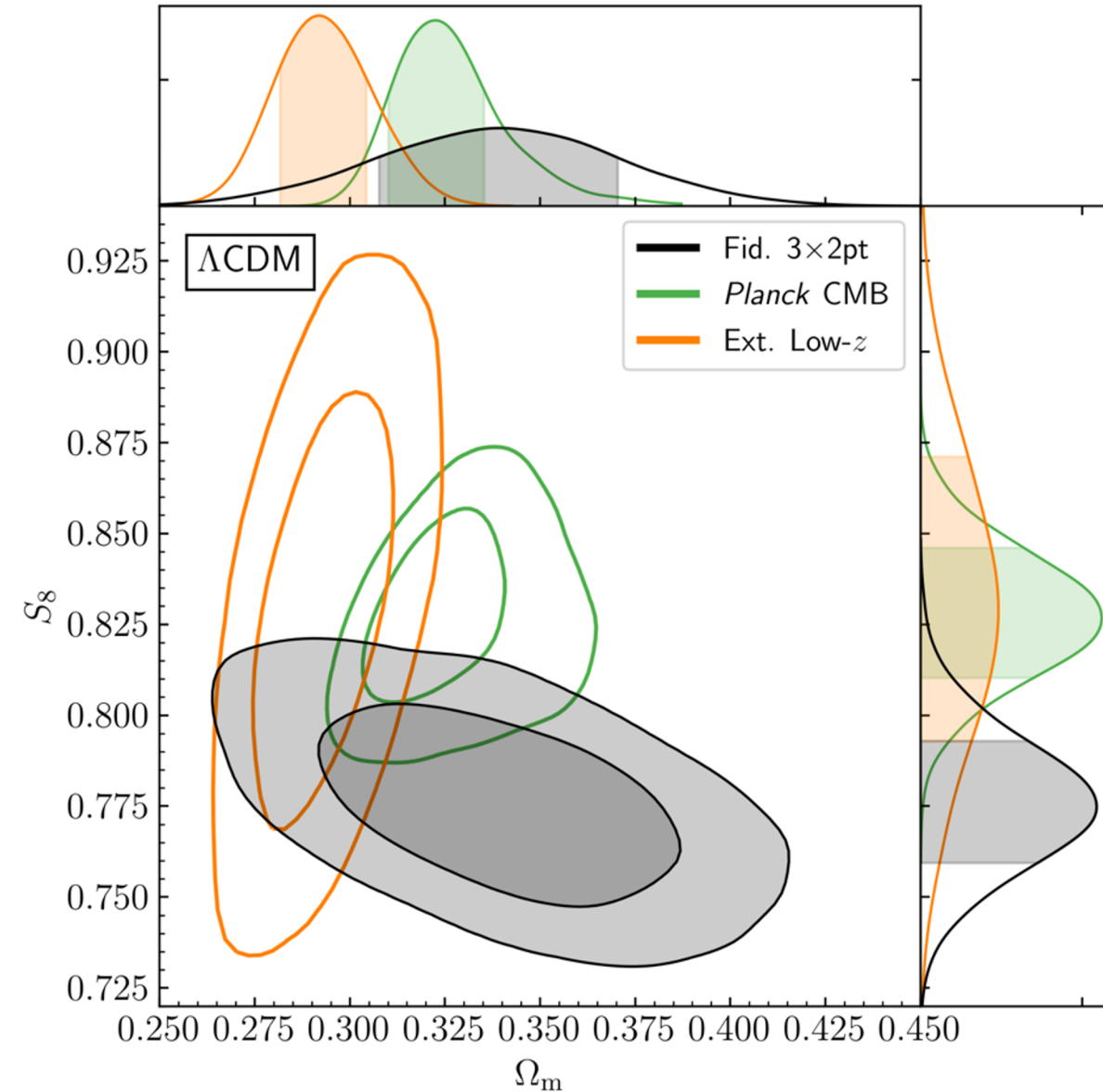
$$w = -0.98^{+0.32}_{-0.20} \quad (-1.03)$$



The Results: Consistency in Λ CDM

We construct three independent data sets:

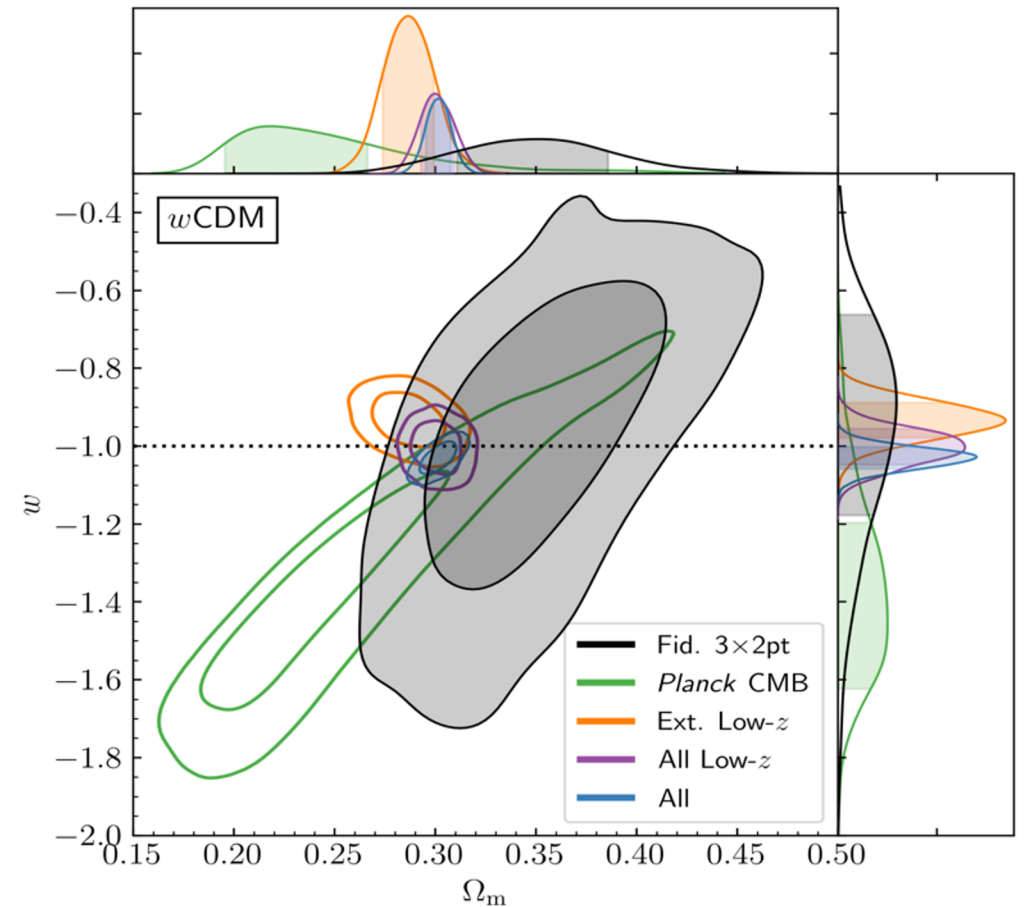
1. Weak lensing and clustering from DES (**3x2pt**)
 2. The combination of other low-redshift **non-lensing** data (**Ext. Low-z**): SNe Ia, BAO, RSD.
 3. **Planck CMB**
- No significant evidence of inconsistency between **DES Y3 3x2pt** and **Planck CMB** ($0.7-1.5\sigma$ or $p=0.13-0.48$)



The Results: Dark Energy

- 3% constraint on w when combined with external datasets
- No evidence for $w \neq -1$ (i.e. consistent with cosmological constant)

$$\begin{aligned}\sigma_8 &= 0.810^{+0.010}_{-0.009} \quad (0.804), \\ \Omega_m &= 0.302^{+0.006}_{-0.006} \quad (0.298), \\ w &= -1.03^{+0.03}_{-0.03} \quad (-1.00)\end{aligned}$$



Open Questions....

- We found indications of systematic effects in some of the lens redshift bins. More work required on sample selection and characterization for **angular** clustering analyses
- Meanwhile the weak lensing constraints are significantly degraded by our conservative intrinsic alignment modeling choice – can we use priors from simulations or external data to improve on this for Y6 / LSST?
- How much can going beyond 2-point statistics improve constraining power to further test Λ CDM? Some papers exploring this are in prep.

Papers available at: <https://www.darkenergysurvey.org/des-year-3-cosmology-results-papers/>

Dark Energy Survey Year 3 results. List of key and supporting papers

1. “Blinding Multi-probe Cosmological Experiments” J. Muir, G. M. Bernstein, D. Huterer et al., arXiv: 1911.05929, MNRAS **494** (2020) 4454
2. “Photometric Data Set for Cosmology”, I. Sevilla-Noarbe, K. Bechtol, M. Carrasco Kind et al., arXiv:2011.03407, ApJS **254** (2021) 24
3. “Weak Lensing Shape Catalogue”, M. Gatti, E. Sheldon, A. Amon et al., arXiv:2011.03408, MNRAS **504** (2021) 4312
4. “Point Spread Function Modelling”, M. Jarvis, G. M. Bernstein, A. Amon et al., arXiv:2011.03409, MNRAS **501** (2021) 1282
5. “Measuring the Survey Transfer Function with Balrog”, S. Everett, B. Yanny, N. Kuropatkin et al., arXiv:2012.12825
6. “Deep Field Optical + Near-Infrared Images and Catalogue”, W. Hartley, A. Choi, A. Amon et al., arXiv:2012.12824
7. “Blending Shear and Redshift Biases in Image Simulations”, N. MacCrann, M. R. Becker, J. McCullough et al., arXiv:2012.08567
8. “Redshift Calibration of the Weak Lensing Source Galaxies”, J. Myles, A. Alarcon, A. Amon et al., arXiv:2012.08566
9. “Redshift Calibration of the MagLim Lens Sample using Self-Organizing Maps and Clustering Redshifts”, G. Giannini et al., in prep.
10. “Clustering Redshifts – Calibration of the Weak Lensing Source Redshift Distributions with redMaGiC and BOSS/eBOSS”, M. Gatti, G. Giannini, et al., arXiv:2012.08569
11. “Calibration of Lens Sample Redshift Distributions using Clustering Redshifts with BOSS/eBOSS”, R. Cawthon et al. arXiv:2012.12826
12. “Phenotypic Redshifts with SOMs: a Novel Method to Characterize Redshift Distributions of Source Galaxies for Weak Lensing Analysis” R. Buchs, C. Davis, D. Gruen et al. arXiv:1901.05005, MNRAS **489** (2019) 820
13. “Marginalising over Redshift Distribution Uncertainty in Weak Lensing Experiments”, J. Cordero, I. Harrison et al., in prep.
14. “Exploiting Small-Scale Information using Lensing Ratios”, C. Sánchez, J. Prat et al., in prep.
15. “Cosmology from Combined Galaxy Clustering and Lensing - Validation on Cosmological Simulations”, J. de Rose et al., in prep.
16. “Unbiased fast sampling of cosmological posterior distributions”, P. Lemos, R. Rollins, N. Weaverdyck, A. Ferte, A. Liddle et al., in prep.
17. “Assessing Tension Metrics with DES and Planck Data”, P. Lemos, M. Raveri, A. Campos et al., arXiv:2012.09554
18. “Dark Energy Survey Internal Consistency Tests of the Joint Cosmological Probe Analysis with Posterior Predictive Distributions”, C. Doux, E. Baxter, P. Lemos et al. arXiv:2011.03410, MNRAS **503** (2021) 2688
19. “Covariance Modelling and its Impact on Parameter Estimation and Quality of Fit”, O. Friedrich, F. Andrade-Oliveira, H. Camacho et al., arXiv:2012.08568
20. “Multi-Probe Modeling Strategy and Validation”, E. Krause et al., in prep.
21. “Curved-Sky Weak Lensing Map Reconstruction”, N. Jeffrey, M. Gatti, C. Chang et al., in prep.
22. “Galaxy Clustering and Systematics Treatment for Lens Galaxy Samples”, M. Rodríguez-Monroy, N. Weaverdyck, J. Elvin-Poole, M. Crocce et al., in prep.
23. “Optimizing the Lens Sample in Combined Galaxy Clustering and Galaxy-Galaxy Lensing Analysis”, A. Porredon, M. Crocce et al., arXiv:2011.03411 PhRvD **103** (2021) 043503
24. “High-Precision Measurement and Modeling of Galaxy-Galaxy Lensing”, J. Prat, J. Blazek, C. Sánchez et al., in prep.
25. “Constraints on Cosmological Parameters and Galaxy Bias Models from Galaxy Clustering and Galaxy-Galaxy Lensing using the redMaGiC Sample”, S. Pandey et al., in prep.
26. “Cosmological Constraints from Galaxy Clustering and Galaxy-Galaxy Lensing using the Maglim Lens Sample” A. Porredon, M. Crocce et al., in prep.
27. “Cosmology from Cosmic Shear and Robustness to Data Calibration”, A. Amon, D. Gruen, M. A. Troxel et al., in prep.
28. “Cosmology from Cosmic Shear and Robustness to Modeling Assumptions”, L. Secco, S. Samuroff et al., in prep.
29. “Magnification modeling and impact on cosmological constraints from galaxy clustering and galaxy-galaxy lensing”, J. Elvin-Poole, N. MacCrann et al., in prep.
30. “Cosmological Constraints from Galaxy Clustering and Weak Lensing” The DES Collaboration in prep.

List of participants

(Early Career Scientists in bold)

	Adam Amara	Ramon Miquel	Alyssa Garcia	Felipe Andrade-Oliveira	Ken Herner	Alex Drlica-Wagner
Beatrice Moser	Santiago Avila	Jenna Freudenberg	Dhayaa Anbajagane	Jack Elvin-Poole	Danielle Leonard	Simon Birrer
Dan Scolnic	Sunayana Bhargava	David Bacon	Andresa Campos	Juan P. Cordero	Gaston Gutierrez	Brian Yanny
Robert Morgan	Antonella Palmese	Tomasz Kacprzak	Cyrille Doux	Mike Jarvis	Federica Tarsitano	Sahar Allam
Paul Rogozenski	Zhiyuan Zhou	Giulia Giannini	Jessie Muir	Eric Huff	Juan Mena Fernández	Scott Dodelson
Elisabeth Krause	Aaron Roodman	Chihway Chang	Georgios Zacharegkas	Chris Conselice	David Sánchez Cid	Jim Annis
Joe DeRose	Matthew Becker	Anderson Souza	William Hartley	Eric Neilsen	Seshadri Nadathur	Andras Kovacs
Richard Kron	Risa Wechsler	Jacobo Asorey	Nick Kokron	Javier Sanchez	Gary Bernstein	Hugo Camacho
H. Thomas Diehl	Andrés Plazas	David Burke	Michael Troxel	Andres Navarro	Sujeong Lee	Kai Hoffmann
Ofer Lahav	Rafael Gomes	Isaac Tutusaus	Judit Prat	Tae-hyeon Shin	Prudhvi Varma	Mandeep Gill
Reese Wilkinson	Ian Harrison	Jamie McCullough	Pablo Fosalba	Chun-Hao To	Oliver Friedrich	Jonathan Blazek
Peter Melchior	Romain Buchs	Paul Ricker	Douglas Tucker	Tesla Jeltema	Simon Samuroff	Lucas James Faga
David Weinberg	Ami Choi	Eduardo Roza	Eli Rykoff	Kevin Wang	Richard Kessler	Joe Zuntz
Anqi Chen	Maria Pereira	Noah Weaverdyck	Michael Johnson	Niall MacCrann	Huan Lin	Steve Kent
Dominik Zuercher	Alex Alarcon	Pauline Vielzeuf	Masaya Yamamoto	Erin Sheldon	Rutuparna Das	Martin Crocce
Niall Jeffrey	Bhuvnesh Jain	Eusebio Sanchez	Dillon Brout	Agnes Ferte	Lorne Whiteway	Spencer Everett
Mitch McNanna	Raphael Sgier	Boyan Yin	Matias Carrasco	Ross Cawthon	Anushka Shrivastava	Juan Estrada
Alexandre Refregier	Albert Stebbins	Robert Gruendl	Daniel Gomes	Manda Banerji	Tamara Davis	Donald Petravick
Dylan Britt	Dragan Huterer	Vivian Miranda	Nico Hamaus	Yuuki Omori	Jimena Gonzalez	Hung-Jin Huang
Pablo Lemos	Justin Myles	Xiao Fang	Ismael Ferrero	Brenna Flaugher	Tim Eifler	Yuanyuan Zhang
Alexandra Amon	Youngsoo Park	Marco Gatti	Mike Wang	Alfredo Zenteno	Giorgia Pollina	Georgios Zacharegkas
Shantanu Desai	Marco Raveri	Heidi Wu	Andrew Liddle	Mathew Smith	Ashley Ross	Shivam Pandey
			Daniel Gruen	Otavio Alves	Eleonora di Valentino	Helen Qu
			Keith Bechtol	Eve Kovacs	Lucas Secco	Eric Baxter
			Juan De Vicente	Martin Rodriguez Monroy	Ji Won Park	Jack Odonnell
			Anna Porredon	Megan Tabbutt	Andrew Pace	Sebastian Bocquet

